Bounding distances to unsafe sets

Jared Miller, Mario Sznaier Supplemental Presentation



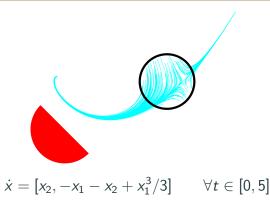
Main Ideas

Quantify safety of trajectories by distance to unsafe set

Relax distance using optimal transport theory

Develop occupation measure programs to bound distance

Flow System Setting



$$X_0 = \{x \mid (x_1 - 1.5)^2 + x_2 \le 0.4^2\}$$

$$X_u = \{x \mid x_1^2 + (x_2 + 0.7)^2 \le 0.5^2,$$

$$\sqrt{2}/2(x_1 + x_2 - 0.7) \le 0\}$$

Distance Function

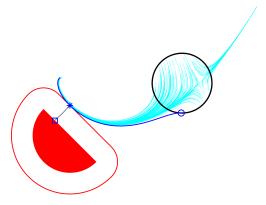
Metric space (X, c) satisfying $\forall x, y \in X$:

$$c(x,y) > 0$$
 $x \neq y$
 $c(x,x) = 0$
 $c(x,y) = c(y,x)$
 $c(x,y) \leq c(x,z) + c(z,y)$ $\forall z \in X$

Point-Unsafe Set distance: $c(x; X_u) = \min_{y \in X_u} c(x, y)$

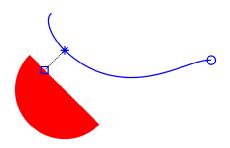
Distance Estimation Problem

$$P^* = \min_{t, x_0 \in X_0} c(x(t \mid x_0); X_u)$$
$$\dot{x}(t) = f(t, x), \quad \forall t \in [0, T].$$



L₂ bound of 0.2831

Optimal Trajectories (Distance)



Optimal trajectories described by $(x_p^*, y^* x_0^*, t_p^*)$:

 x_p^* location on trajectory of closest approach

y* location on unsafe set of closest approach

 x_0^* initial condition to produce x_p^*

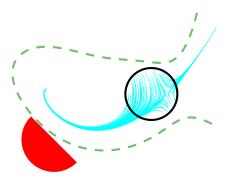
 t_p^* time to reach x_p^* from x_0^*

Safety Background

Barrier Program

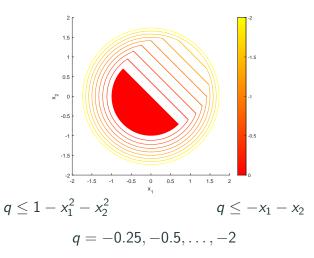
Barrier function $B: X \to \mathbb{R}$ indicates safety

$$B(x) \le 0$$
 $\forall x \in X_u$
 $B(x) > 0$ $\forall x \in X_0$
 $\mathcal{L}_f B(x) \ge 0$ $\forall x \in X$



Half-circle Contours

Unsafe set $X_u = \{x \mid 1 - x_1^2 - x_2^2 \ge 0, -x_1 - x_2 \ge 0\}$

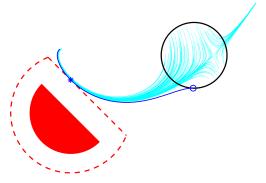


Safety Margin

Unsafe set $X_u = \{x \mid p_i(x) \geq 0 \ \forall i = 1 \dots N_u\}$

Safety margin $p^* = \max \min_i p_i(x)$ along trajectories

If $p^* < 0$, no trajectories enter X_u (safe)



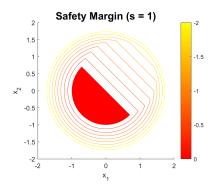
safe: $p^* \le -0.2831$

Safety Margin Scaling

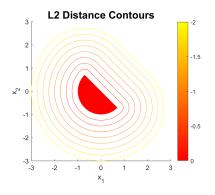
Scale factor in constraints

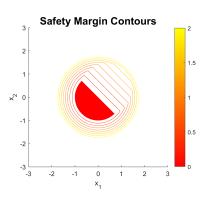
$$q \le 1 - x_1^2 - x_2^2$$

$$q \leq s(-x_1-x_2)$$



Distance vs. Safety Margin



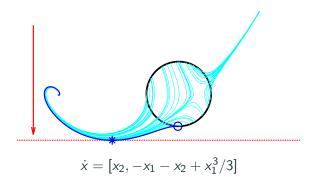


Peak Estimation

Peak Estimation Background

Find maximum value of p(x) along trajectories

$$P^* = \max_{t, x_0 \in X_0} p(x(t \mid x_0))$$
$$\dot{x}(t) = f(t, x(t)) \qquad t \in [0, T]$$



Peak Estimation Program (Measure)

Infinite dimensional linear program (Cho, Stockbridge, 2002)

$$p^* = \max \quad \langle p(x), \mu_p \rangle$$
 $\mu_p = \delta_{t=0} \otimes \mu_0 + \mathcal{L}_f^{\dagger} \mu$
 $\langle 1, \mu_0 \rangle = 1$
 $\mu, \mu_p \in \mathcal{M}_+([0, T] \times X)$
 $\mu_0 \in \mathcal{M}_+(X_0)$

Peak measure μ_p : free terminal time

Maximin Program

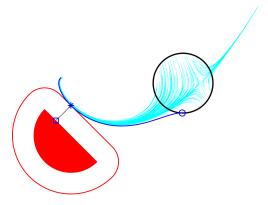
Solve
$$P^* = \max_x \min_i p_i(x)$$
 (M., Henrion, Sznaier 2020)
$$p^* = \max_{\boldsymbol{q}} \boldsymbol{q}$$
 $\boldsymbol{q} \leq \langle p_i(x), \mu_p \rangle$ $\forall i$
$$\mu_p = \delta_{t=0} \otimes \mu_0 + \mathcal{L}_f^\dagger \mu$$
 $\langle 1, \mu_0 \rangle = 1$
$$\mu, \mu_p \in \mathcal{M}_+([0, T] \times X)$$
 $\mu_0 \in \mathcal{M}_+(X_0)$

Used for safety margins, $p^* \le p_d^* < 0$

Distance Program

Distance Estimation Problem (reprise)

$$P^* = \min_{t, x_0 \in X_0} c(x(t \mid x_0); X_u)$$
$$\dot{x}(t) = f(t, x), \quad \forall t \in [0, T].$$



L₂ bound of 0.2831

Distance Relaxation

Distance in points \rightarrow Earth-Mover distance

$$c(x,y) \qquad \langle c(x,y), \eta \rangle x \in X \quad \to \quad \langle 1, \eta \rangle = 1 y \in X_u \qquad \eta \in \mathcal{M}_+(X \times X_u)$$

Joint (Wasserstein) probability measure η

Measures from Optimal Trajectories

Form measures from each $(x_p^*, x_0^*, t_p^*, y^*)$

Atomic Measures (rank-1)

$$\mu_0^*$$
: $\delta_{x=x_0^*}$

$$\mu_p^*$$
: $\delta_{t=t_p^*} \otimes \delta_{x=x_p^*}$

$$\eta^*$$
: $\delta_{x=x_p^*}\otimes\delta_{y=y^*}$

Occupation Measure
$$\forall v(t,x) \in C([0,T] \times X)$$

$$\mu^*$$
: $\langle v(t,x),\mu\rangle = \int_0^{t_p^*} v(t,x^*(t\mid x_0^*))dt$

Distance Program (Measures)

Infinite Dimensional Linear Program (Convergent)

$$p^* = \min \quad \langle c(x, y), \eta \rangle$$

$$\pi_\#^{\times} \eta = \pi_\#^{\times} \mu_p$$

$$\mu_p = \delta_0 \otimes \mu_0 + \mathcal{L}_f^{\dagger} \mu$$

$$\langle 1, \mu_0 \rangle = 1$$

$$\eta \in \mathcal{M}_+(X \times X_u)$$

$$\mu_p, \ \mu \in \mathcal{M}_+([0, T] \times X)$$

$$\mu_0 \in \mathcal{M}_+(X_0)$$

Prob. Measures:
$$\langle 1, \mu_0 \rangle = \langle 1, \mu_p \rangle = \langle 1, \eta \rangle = 1$$

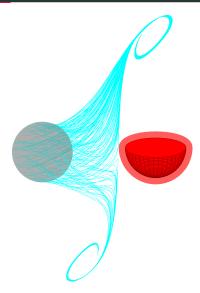
Distance Example (Twist)

'Twist' System,
$$T = 5$$

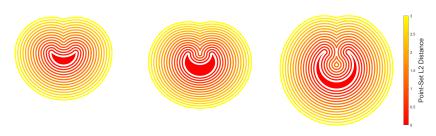
$$\dot{x}_i = A_{ij}x_j - B_{ij}(4x_j^3 - 3x_j)/2$$

$$A = \begin{bmatrix} -1 & 1 & 1 \\ -1 & 0 & -1 \\ 0 & 1 & -2 \end{bmatrix}$$

$$B = \begin{vmatrix} -1 & 0 & -1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{vmatrix}$$

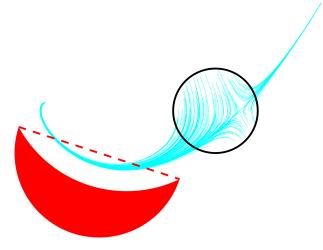


Moon L2 Contours



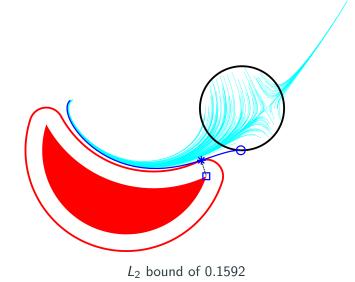
Inside one circle, outside another

Distance Example (Flow Moon)



Collision if X_u was a half-circle

Distance Example (Flow Moon)



Distance Variations

Sparsity

Separable
$$c(x, y) = \sum_i c_i(x_i, y_i)$$

Use correlative sparsity with measures and cliques

$$\eta_k$$
: $I_k = (x_k : x_n, y_1 : y_k)$ $\forall k = 1, \ldots, n$

Sparse decomposition of η :

$$\min \sum_{i} \langle c_i(x_i, y_i), \eta_i \rangle \qquad \eta^1 \in \mathcal{M}_+(X \times \mathbb{R})
\pi_\#^{I_k \cap I_{k+1}} \eta_k = \pi_\#^{I_k \cap I_{k+1}} \eta_{k+1} \qquad \eta^k \in \mathcal{M}_+(\mathbb{R}^{n+1})
\pi_\#^{\times} \mu^p = \pi_\#^{\times} \eta_1 \qquad \eta^n \in \mathcal{M}_+(\mathbb{R} \times X_u)$$

Shapes along Trajectories

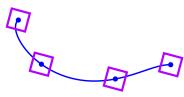
Orientation $\omega(t) \in \Omega$, shape S

Body to global coordinate transformation A:

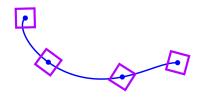
$$A: S \times \Omega \rightarrow X$$

$$(s,\omega)\mapsto A(s;\omega)$$

Angular Velocity = 0 rad/sec



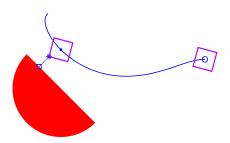
Angular Velocity = 1 rad/sec



Set-Set Distance Problem

Set-Set distance between $A(\cdot; \omega) \circ S$ and X_u given ω

$$P^* = \min_{t, \omega_0 \in \Omega_0, s \in S} c(x(t); X_u)$$
$$x(t) = A(s; \omega(t \mid \omega_0)) \quad \forall t \in [0, T]$$
$$\dot{\omega}(t) = f(t, \omega) \qquad \forall t \in [0, T]$$



L₂ bound of 0.1465

Set-Set Program (Measures)

Add new 'shape' measure μ_s

$$p^* = \min \quad \langle c(x, y), \eta \rangle$$

$$\mu_p = \delta_0 \otimes \mu_0 + \mathcal{L}_f^{\dagger} \mu$$

$$\pi_\#^{\omega} \mu_p = \pi_\#^{\omega} \mu_s$$

$$\pi_\#^{\chi} \eta = A(s; \omega)_\# \mu_s$$

$$\langle 1, \mu_0 \rangle = 1$$

$$\eta \in \mathcal{M}_+(X \times X_u)$$

$$\mu_s \in \mathcal{M}_+(\Omega \times S)$$

$$\mu_p, \ \mu \in \mathcal{M}_+([0, T] \times \Omega)$$

$$\mu_0 \in \mathcal{M}_+(\Omega_0)$$

Distance Variations

Uncertainty in dynamics

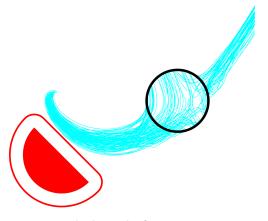
Lifted distances (with absolute values)

Set-Set distances for shape safety

Distance Uncertainty

Time dependent uncertainty $w(t) \in W \ \forall t \in [0, T]$

Young measure μ , Liouville $\mu_p = \delta_0 \otimes \mu_0 + \pi_\#^{tx} \mathcal{L}_f^\dagger \mu$



L₂ bound of 0.1691

Lifted Distance



LP lifts to deal with absolute values

$$||x - y||_{\infty}$$
 min q

$$-q \le \langle x_i - y_i, \eta \rangle \le q \qquad \forall i$$

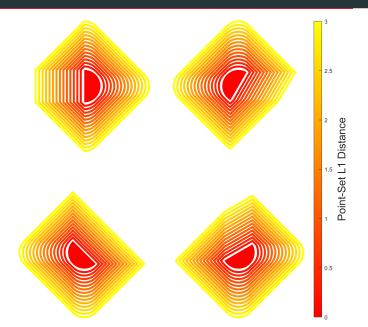


$$||x - y||_1$$
 min $\sum_i q_i$
 $-q_i \le \langle x_i - y_i, \eta \rangle \le q_i$ $\forall i$

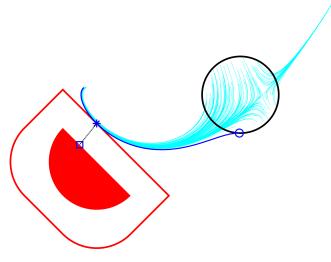


$$\|x - y\|_3^3$$
 min $\sum_i q_i$
 $-q_i \le \langle (x_i - y_i)^3, \eta \rangle \le q_i \quad \forall i$

Half-Circle L1 Contours



Lifted Distance (L1) Example



 L_1 bound of 0.4003

Take-aways

Conclusion

Distance Estimation with occupation measures

Approximate recovery if moment matrices are low-rank

Extend to uncertain, lifted, set-set scenarios

Future Work

- Distance-Maximizing Control
- Further Sparsity
- Efficient Computation

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Thank you

Thank you for your attention